
Foundations of Natural Language Processing

Lecture 19c

Constructing Representations of Discourse Coherence

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So Far

- Understanding discourse involves identifying the **coherence relations**
 - Narration, Explanation, Background, Contrast, Parallel, QA, Correction. . .
- Discourse coherence influences many pragmatic phenomena.
- LFs that feature coherence relations can be formally interpreted

Now: How to automatically infer coherence relations

Recap: an Example

π_1 : John can open Bill's safe.

π_2 : He knows the combination.

π_0 : *Explanation*(π_1, π_2)

π_1 : $\iota x(\mathit{safe}(x) \wedge \mathit{possess}(x, \mathit{bill}) \wedge \mathit{can}(\mathit{open}(e_1, \mathit{john}, x)))$

π_2 : $\iota y(\mathit{combination}(y) \wedge \mathit{of}(y, x) \wedge \mathit{knows}(\mathit{john}, y))$

- Bits in **red** are specific values that go beyond content that's revealed by linguistic form.
- They are inferred via **commonsense reasoning** that's used to construct a **maximally coherent** interpretation.

Symbolic approaches to constructing LF

- Draw on rich information sources:
 - linguistic content, world knowledge, mental states. . .
- Deploy reasoning that supports inference with partial information. Unlike classical logic, this requires **consistency tests**.
- Typically, construct LF and evaluate it **in the same logic**, making constructing LF **undecidable**.

Further Problem

- Like any knowledge rich approach involving hand-crafted rules, this is only feasible for very small domains.
- Ideally, we would like to **learn** a discourse parser automatically from corpus data.
- But there's a lack of corpora annotated with discourse structure.
 - RSTbank, Graphbank, Annodis, STAC are relatively small.
 - Discourse Penn Treebank is relatively large but not annotated with complete discourse structure.
 - Groningen Parallel Meaning Bank: full discourse structure (SDRSs) and getting bigger all the time.

Supervised Learning for SDRT

Training on 100 dialogues

Baldrige and Lascarides (2005)

Parser based on Collins' parsing model:

- 72% f-score on segmentation (baseline: 53.3%)
- 48% f-score on segmentation and coherence relations (baseline: 7.4%)
- Doesn't attempt to estimate LFs of clauses.

Training on Groningen Meaning Bank

Liu and Lapata (2018)

Neural semantic parser, RNN computes structure first, fills in arguments later:

- 77% f-score on segmentation, coherence relations *and* LFs of clauses
- State of the Art!

- Coherence relations can be overtly signalled:
 - *because* signals EXPLANATION; *but* signals CONTRAST
- So produce a training set *automatically*:
 - Max fell because John pushed him
⇒
EXPLANATION(*Max fell, John pushed him*).

Results of Best Model

- Test examples originally had a cue phrase: 60.9%.
- Test examples originally had **no cue phrase**: 25.8%
- Train on 1K manually labelled examples: 40.3%.
- Combined training set of manual and automatically labelled examples doesn't improve accuracy.

So you're better off manually labelling a small set of examples!

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Why?

Contrast to Elaboration

Although the electronics industry has changed greatly, possibly the greatest change is that very little component level manufacture is done in this country.

Summary

- Interpretation governed by discourse coherence:
 - Constrains what can be said next
 - Augments meaning revealed by linguistic form.
- Computing logical form should be decidable; modularity is key to this.
- Data-driven approaches are a major challenge.
- Linking rich models of discourse semantics to models of human behaviour and decision making is also a major challenge, but essential for tackling dialogues where the agents' goals conflict.